



A Gaussian Derivative Model of the Complex Cell

Miles Hansard, Radu Horaud

► To cite this version:

Miles Hansard, Radu Horaud. A Gaussian Derivative Model of the Complex Cell. ECVP 2010 - 33rd European Conference on Visual Perception, Aug 2010, Lausanne, Switzerland. 2010. inria-00590263

HAL Id: inria-00590263

<https://inria.hal.science/inria-00590263>

Submitted on 3 May 2011

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

1. Introduction

- Visual filters can be modelled by derivatives G_k of the Gaussian function.
- This **Gaussian jet** representation is convenient because it is:
 - Steerable, hence all orientations can be represented concisely.
 - Dimensionally separable, hence easily defined in 2D and 3D.
 - The natural code for typical image features, e.g. edges and blobs.
- But what about **complex cells**, cf. the Gabor energy model?
- Can the jet be made insensitive to small **shifts of the image**?

2. Subunit Filters

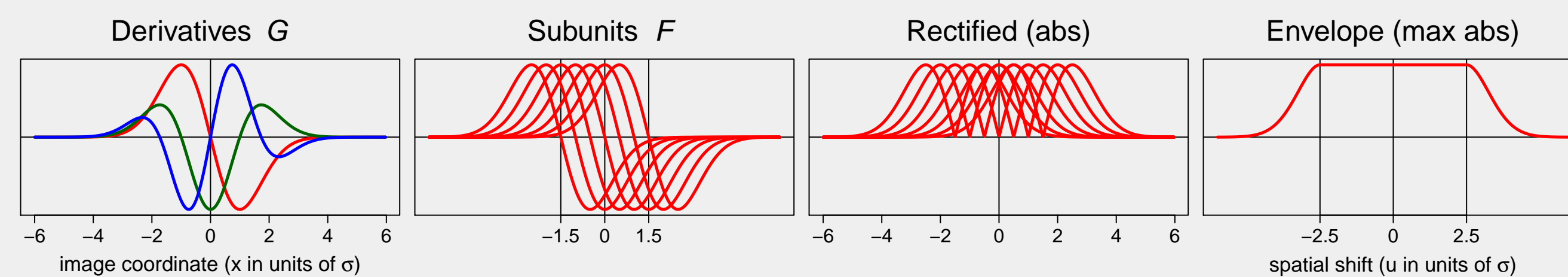
- Let $F_\star(x, u)$ be a family of **subunit filters**, parameterized by shift u .
- These can be Taylor-approximated from $F_\star(x, 0)$ and its derivatives.
- In particular, choose the edge-filters $F_\star(x, u) = G_1(x - u, \sigma)$.
- Approximate filters $F \approx F_\star$ can be obtained from the D th-order jet:

$$F(x, u) = \sum_{k=0}^{D-1} \frac{(-u)^k}{k!} G_{k+1}(x, \sigma)$$

- Problems: Unstable, and nature of the approximation is unclear.
- Solution: Allow **polynomial weights** $P_k(u)$, and solve by least-squares.

3. Invariant Response

- The basis G is used to synthesize the filters F_i , for each shift u_i .
- Each subunit filter is applied to the signal s , giving $q_i = F_i \cdot s$.
- The response-envelope is computed by the operation $\max_i |q_i|$.



- More derivatives are needed in practice (see box 6).
- The range of shifts must cover $\pm\sigma$ for a unimodal impulse-response.
- Note that the family of subunit filters is continuous (only 7 shown).

4. Neural Implementation

- A neurally plausible 'softmax' is used to compute the response-envelope:

$$\max_i |q_i| \approx \sum_i w_i |q_i|$$

- The weights w_i are defined by a **nonlinearity** and **normalization**:

$$w_i = \exp(\mu |q_i|) / \sum_i \exp(\mu |q_i|)$$

5. Matrix Formulation

- F : Subunit filters (rows) P : Polynomials (columns)
- G : Gaussian derivatives (rows) M : Monomials (columns)
- s : Input signal (column) C : Estimated coefficients

- The subunit-filters are polynomially-weighted Gaussian derivatives:

$$F = PG \quad \text{where} \quad P = MC$$

- The filter-design problem is to estimate C , given ideal filters F_\star .

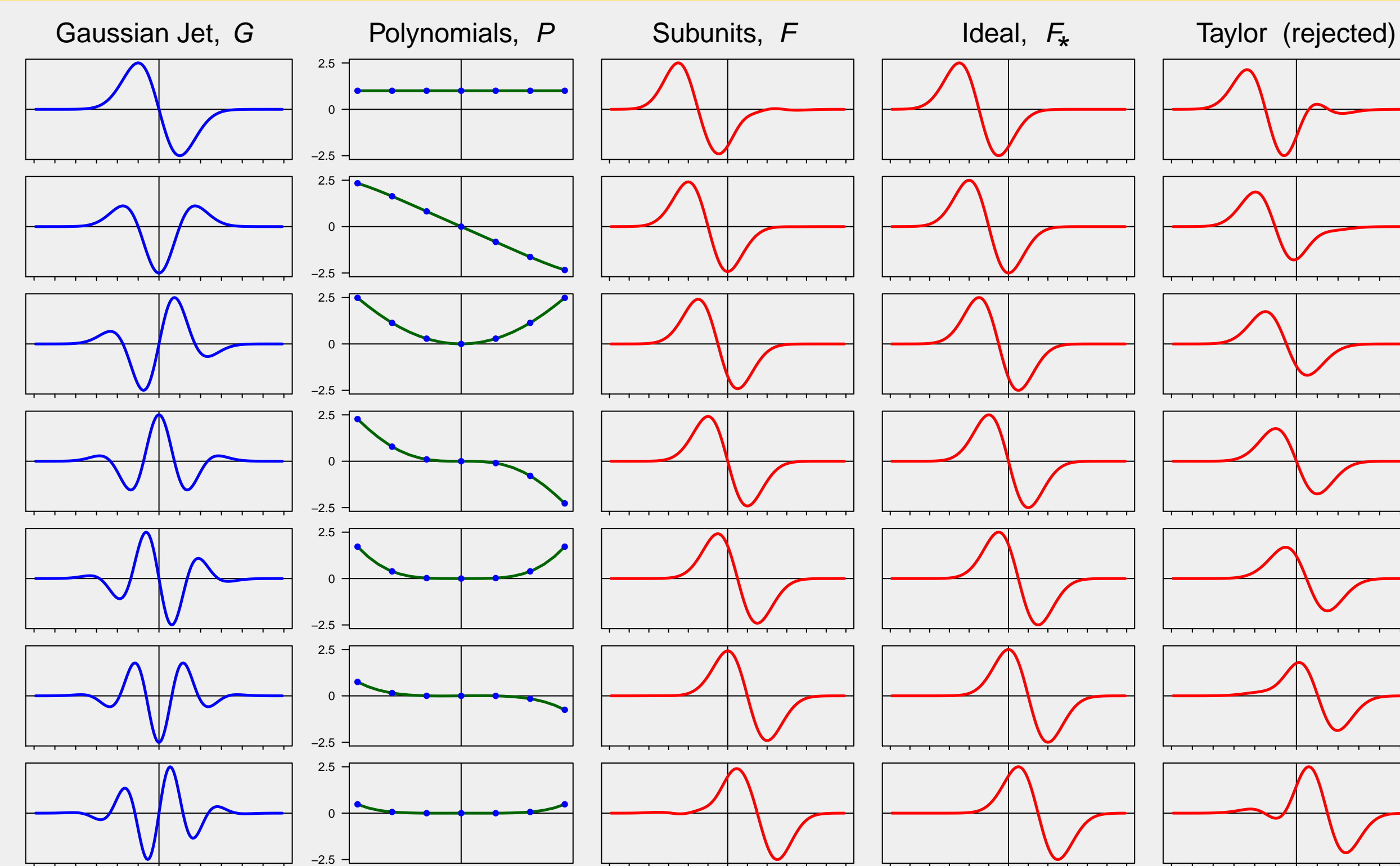
- The **least-squares solution** involves two pseudo-inverses:

$$F = MCG \quad \text{where} \quad C = M^+ F_\star^+$$

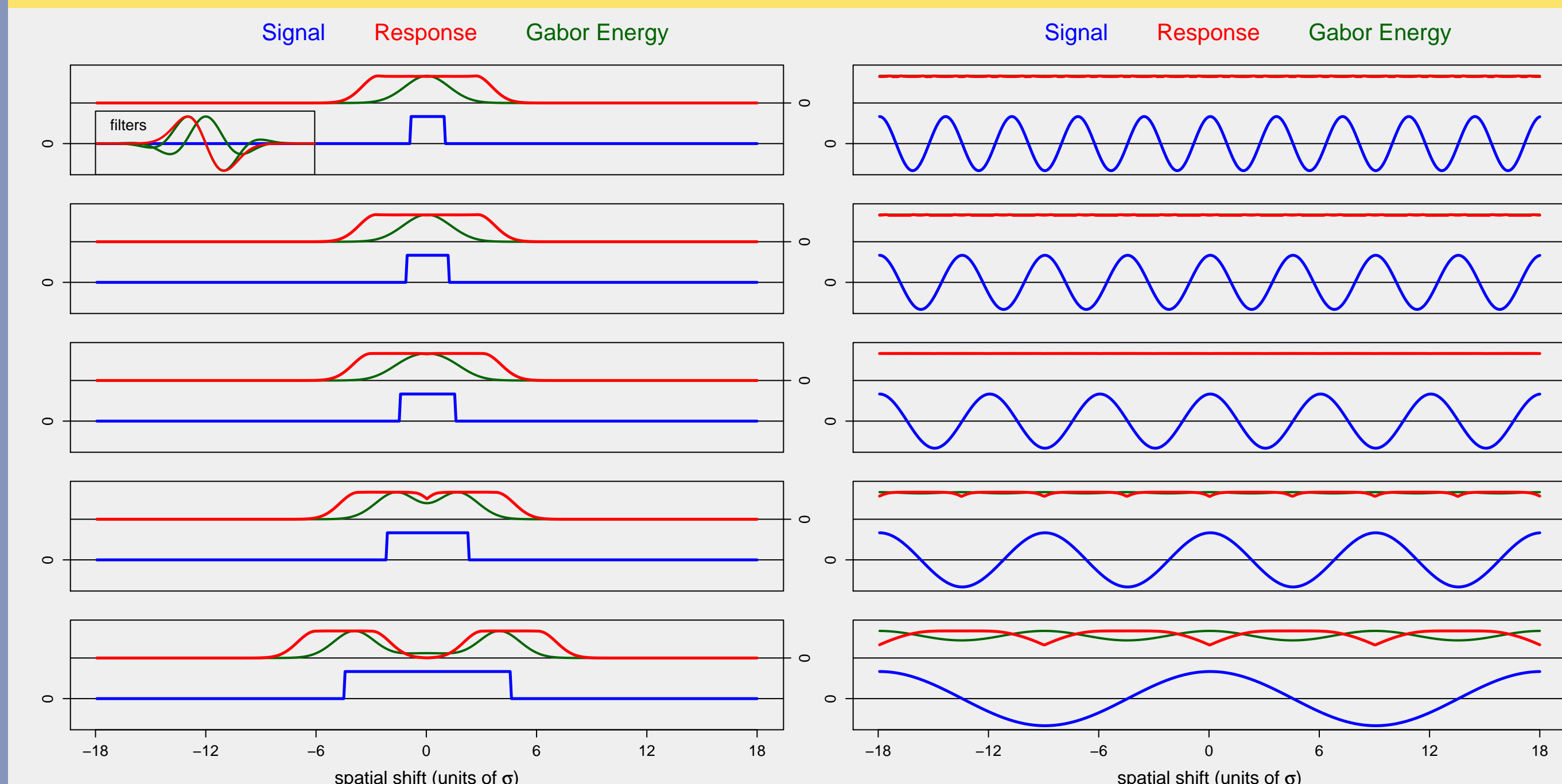
- The subunit response q is a **linear transformation** of the jet response:

$$q = Fs = P(Gs)$$

6. One-Dimensional Example



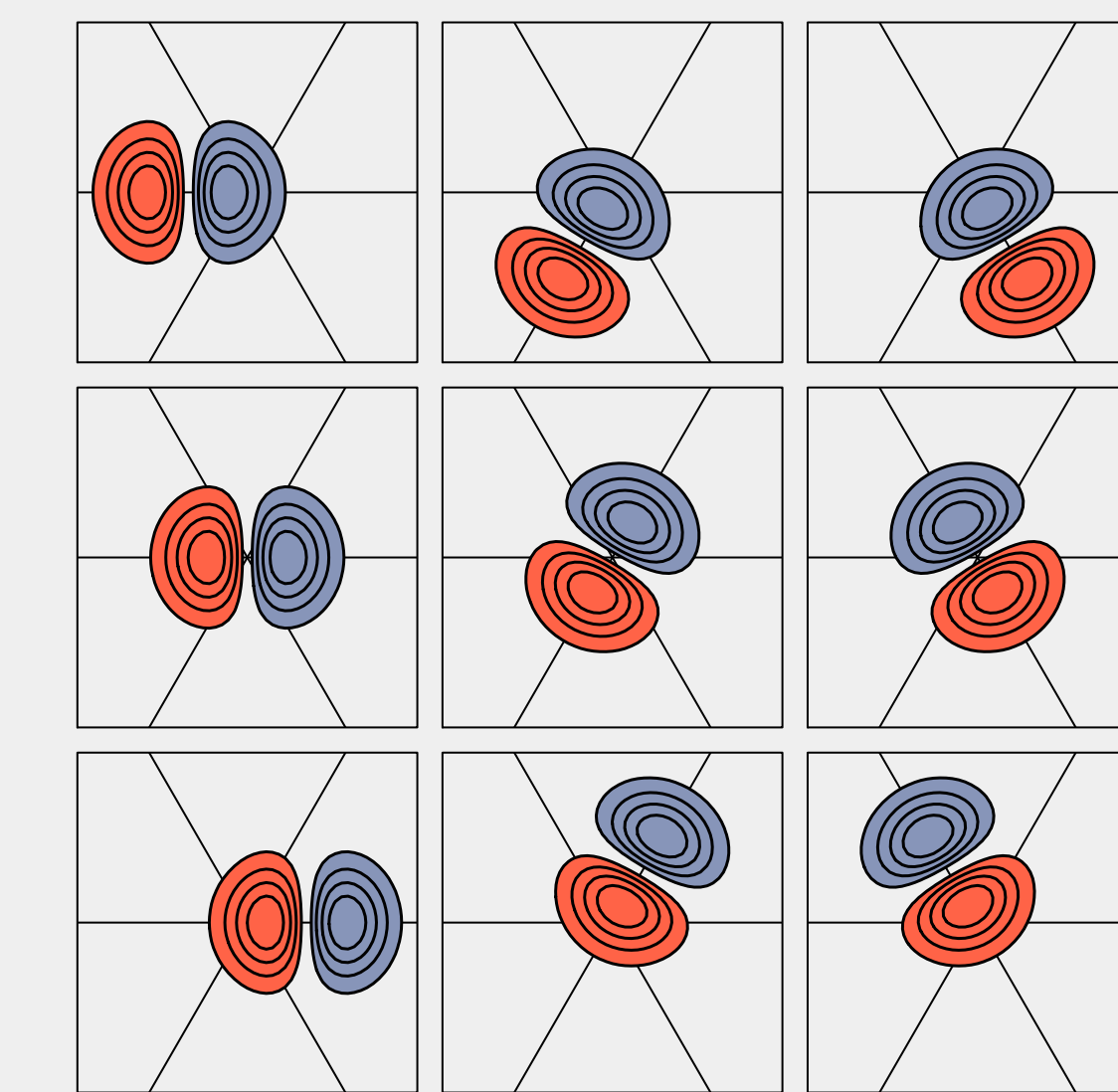
7. Bar and Grating Responses



8. Steerability

- Any Gaussian derivative $G_k(x, \sigma, \theta)$ can be **exactly** synthesized from the basis $G_k(x, \sigma, \theta_j)$, where $j = 1 \dots k + 1$.
- A complete basis of order D requires $\sum_{k=1}^D (k + 1) = \frac{1}{2}D(D + 3)$ filters.
- The basis matrix G must be extended, but there is otherwise no change to the filter-synthesis algorithm (box 5).

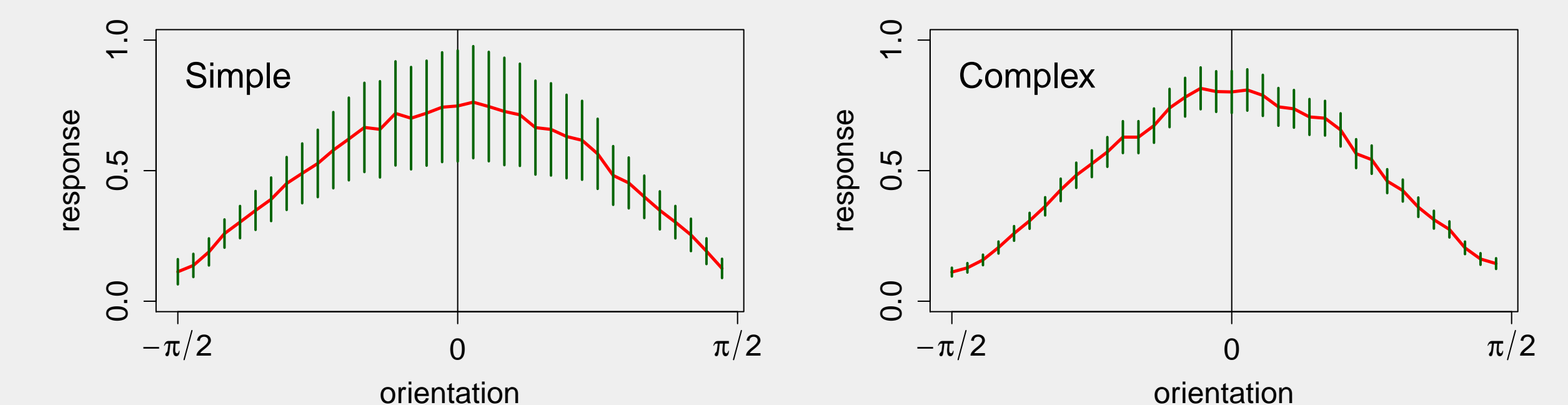
9. Two-Dimensional Example



- The complete basis of order $D = 8$ contains 44 oriented filters, all centred at the **same location**.
- This permits accurate synthesis of **any** filter $G_1(x - u, \sigma, \theta)$, where $u \in \pm 1.5\sigma$ and $\theta \in [0, 2\pi)$.
- Each column corresponds to a complex cell RF, $\theta = 0^\circ, 60^\circ, 120^\circ$.
- Each box shows a subunit.

10. Natural Image Response

- The model is evaluated in the framework of 'slow feature analysis'.
- Pick 100 points at random, from an image with a **dominant orientation**.
- Choose straight 'tracks' at 36 orientations through each point.
- Apply simple/complex cell model at 100 points along each track.
- Compute **mean response** and **SD** in each orientation, over all points.



- The simple response is orientation tuned (red curve), but highly variable.
- The complex response is also orientation tuned, but much less variable.

11. Conclusions

- A shift-insensitive response can be obtained from the Gaussian jet.
- The signal structure can be represented **geometrically**.
- The new model is steerable, and works in any number of dimensions.
- High-order filters, as seen in neural data, are needed in the jet basis.